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import pandas as pd
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
from sklearn.metrics import davies_bouldin_score
customers=pd.read_csv("Customers.csv")
products=pd.read_csv("Products.csv")
transactions = pd.read_csv("Transactions.csv")
merged_data = transactions.merge(customers, on='CustomerID', how='left').merge(products, on='ProductID', how='left')
customer_features = merged_data.groupby('CustomerID').agg({
    'TotalValue': 'sum',
    'Quantity': 'sum',
    'Category': lambda x: x.mode()[0], # Most frequent category
    'Region': 'first'}).reset_index()
# Encode categorical data (Region, Category)
customer_features = pd.get_dummies(customer_features, columns=['Region', 'Category'], drop_first=True)
# K-Means Clustering
kmeans = KMeans(n_clusters=5, random_state=42)
customer_features['Cluster'] = kmeans.fit_predict(customer_features.drop(columns=['CustomerID']))
# Evaluate clustering with Davies-Bouldin Index
db_index = davies_bouldin_score(customer_features.drop(columns=['CustomerID', 'Cluster']), customer_features['Cluster'])
print(f"Davies-Bouldin Index: {db_index}")
Davies-Bouldin Index: 0.5164204673282325
import seaborn as sns
# Visualize clusters
sns.scatterplot(
   x=customer_features['TotalValue'],
    y=customer_features['Quantity'],
    hue=customer_features['Cluster'],
    palette='viridis'
plt.title('Customer Clusters')
plt.show()
→▼
                                   Customer Clusters
              Cluster
         30
                   0
                   1
                   2
         25
                   3
        20
      Quantity
        15
         10
          5
                        2000
                                   4000
                                              6000
                                                         8000
                                                                    10000
                                        TotalValue
```